

In Situ Analysis and Visualization with Catalyst and Ascent

Jean M. Favre, Senior Visualization Software Engineer CSCS

Thank to different teams

- Thanks to the many actors in the visualization and in-situ dev teams:
- Kitware (François Mazen will join me for the tutorial at ISC22)
- Berkeley Lab, ANL, LLNL, Los Alamos, Sandia, and many others (in particular Cyrus Harrison)



Agenda B.C.

13:00	Welcome, Overview and Motivation
	Agenda and technical details for demonstrations
13:15	Introduction to in-situ visualization, workflows and terminology
13:30	Conduit, an API to describe hierarchical scientific data
	The Mesh Blueprint, usage conventions, examples
14:00	Ascent, an in-situ visualization and analysis library using Conduit
	Making images Transforming data, extracting data Queries and Conditional triggers
14:30	Instrumentation of simulation codes with Ascent
JacobiPython: Use jupyter notebooks on Piz Daint SPH-EXA: an open-source parallel particle simulation code LULESH: an open-source parallel unstructured flow solver	





Agenda A.C.

15:30 ParaView Catalyst v2, an in-situ visualization and analysis library using Conduit

The ParaView interactive application, introduction

The Paraview parallel server architecture

The Catalyst API

The ParaView Catalyst Blueprint

Python scripting, Data Extractors

16:00 Instrumentation of simulation codes with Catalyst

SPH-EXA: an open-source parallel particle simulation code

LULESH: an open-source parallel unstructured flow solver

Catalyst, connecting to a live simulation, steering

ParaView Cinema: An Image-Based Approach to Extreme-Scale Data Analysis

16:25 Ascent executing ParaView Python code

16:35 Future developments, alternative workflows (in-transit visualization)

16:45-17:00 Wrap-up, Q&A



Technical details

- course account "class1??" with a password
- ssh ela.cscs.ch + ssh daint.cscs.ch
- https://jupyter.cscs.ch
- Your account has been already bootstrapped with requirements for jupyter lab
 - see \$HOME/.local/shared/jupyter/kernels
- Reservation for computer nodes "in-situ"
- Your account will expire at the end of the day
- My repository should not expire: https://github.com/jfavre/InSitu-Vis-Tutorial2022



Overview

- What is in-situ visualization, why do we need it? What solutions are available to implement it?
- See recent book "In Situ Visualization for Computational Science"
- Some examples from the previous decade (libSim, Catalyst)

Replay some of the presentations from the most recent Ascent tutorial at SC21

Detail the ParaView-Catalyst and Ascent solutions, with practical examples.



"Piz Daint" at CSCS, the Swiss flagship for national HPC Service

- Cray XC40/XC50
- 5704 hybrid nodes (Intel Xeon E5-2690 v3/NVIDIA Tesla P100)
- 1813 multi-core nodes (Intel Xeon E5-2695 v4)







Coming online in 2023, the 'Alps' system infrastructure

will replace CSCS's Piz Daint and serve as a general-purpose system.

Link







The past

- For decades, the dominant paradigm has been post-hoc visualization
- Simulation codes iterate, and save data at regular time intervals.
 - Visualization and domain scientists can then read the data back from storage and interactively explore the data without time constraints
 - In particular at CSCS, see https://user.cscs.ch/computing/visualisation/

"Without I/O, no visualization is possible"

The true cost of doing I/O is an aggregate of the solver's I/O phase and the many iterations of visualization sessions.



Post-hoc visualization

Even if scientists could afford to keep most of the data for analysis, they must transfer the data to a machine with sufficient capacity and processing power:

- → Very high data transfer
- → Visualization machine needs to be almost as powerful as the supercomputer
- → The alternative: use smaller temporal and spatial subsets

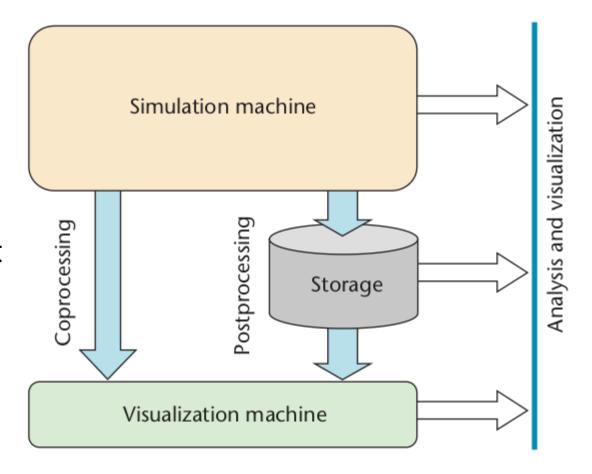


Figure taken from "In Situ Visualization at Extreme Scale: Challenges and Opportunities", Kwan-Liu Ma, IEEE CG&A, nov/dec 2009



in situ visualization

Instrument the code such that both the simulation and visualization calculations run on the same hardware

This runtime co-processing can render images directly or extract features -- which are much smaller than the original raw data

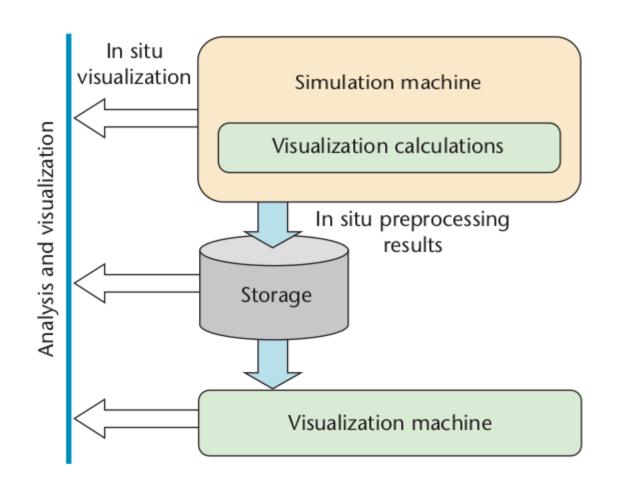


Figure taken from "In Situ Visualization at Extreme Scale: Challenges and Opportunities", Kwan-Liu Ma, IEEE CG&A, nov/dec 2009



In-situ visualization has raised quite a few questions

- Sharing physical resources and domain decomposition?
- What % of time can we afford to "do visualization" vs. "advance the solver"?
- Which feature extraction and visualization tasks are best suited for on-the-fly processing?
- Since less data would be effectively stored to disk, should we augment it with ancillary data?
- Can we provide a generic abstraction to describe the data and mesh structures?



A third paradigm also emerged: in-transit visualization

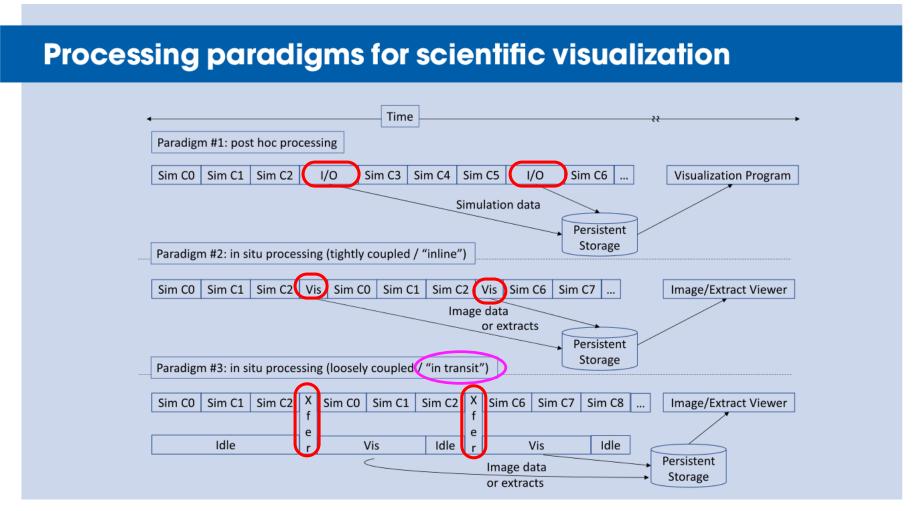


Figure taken from "In Situ Visualization for Computational Science", Hank Childs et al., IEEE CG&A, nov/dec 2019



Does "in-situ" mean "in place"?

The data is already in the processor's memory space, without touching the disks

• If simulation data is moved to a distinct set of resources (nodes dedicated to visualization), we are still analysing data "in place", but is it "in-situ"?

- There has been quite a few variants on the terminology:
 - Co-processing, concurrent processing, run-time visualization



Many definitions and colloquial use for "in-situ"

- An exhaustive panorama of the different systems in use was created :
- "A Terminology for In Situ Visualization and Analysis Systems", Hank Childs et al, International Journal of High Performance Computing Applications, 34(6):676– 691
- http://cdux.cs.uoregon.edu/pubs/ChildsIJHPCA.pdf

• For the scope of this paper, "in situ processing" was defined to be:

"processing data as it is generated"



in situ systems were best described via multiple, distinct axes

integration type

How visualization and analysis code is integrated with the simulation code?

proximity

How close is the visualization code from the data?

Access

How does the simulation give access to the data?

division of execution:

how compute resources are shared between simulation and in situ routines.

operation controls:

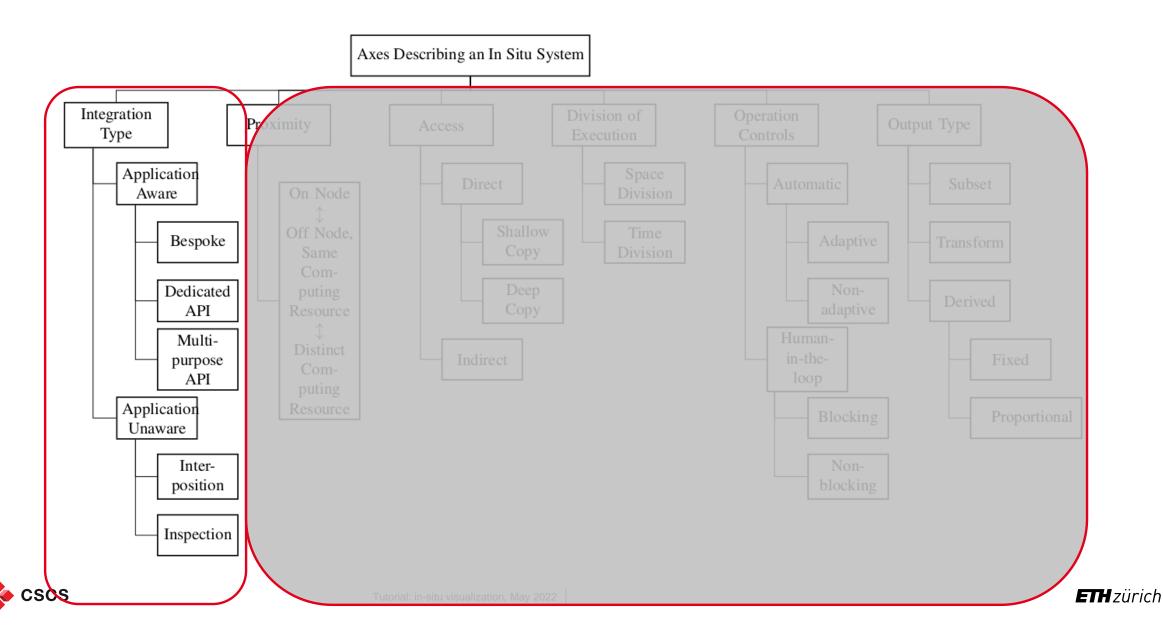
the mechanism for selecting which operations are executed during run-time

output type

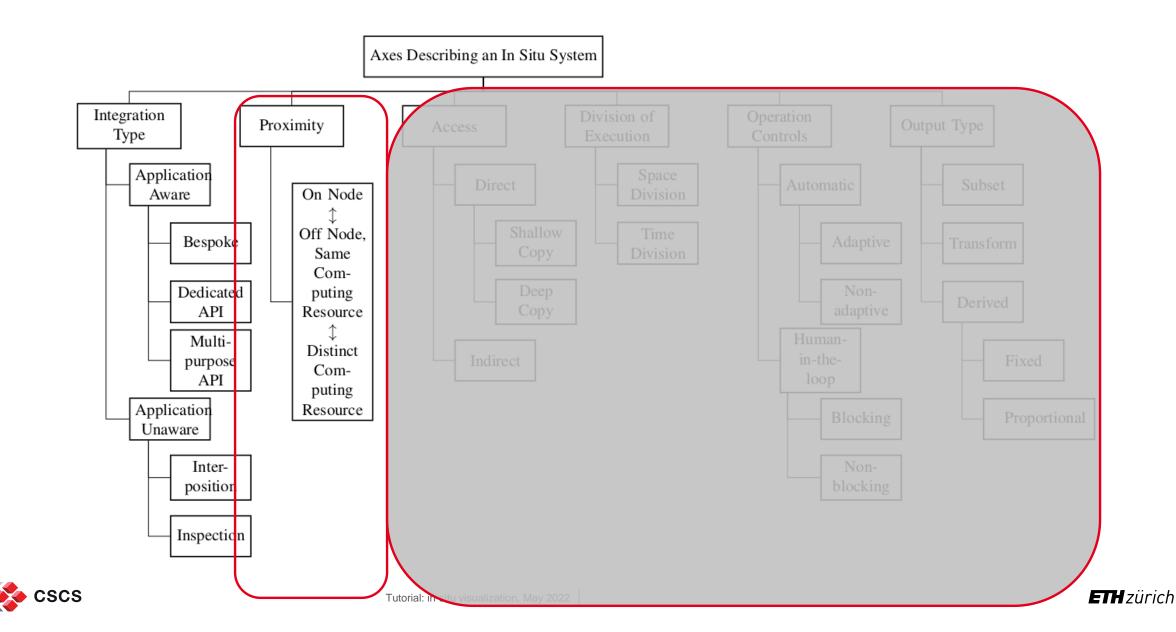
which types of operations are performed on the simulation data before it is output.



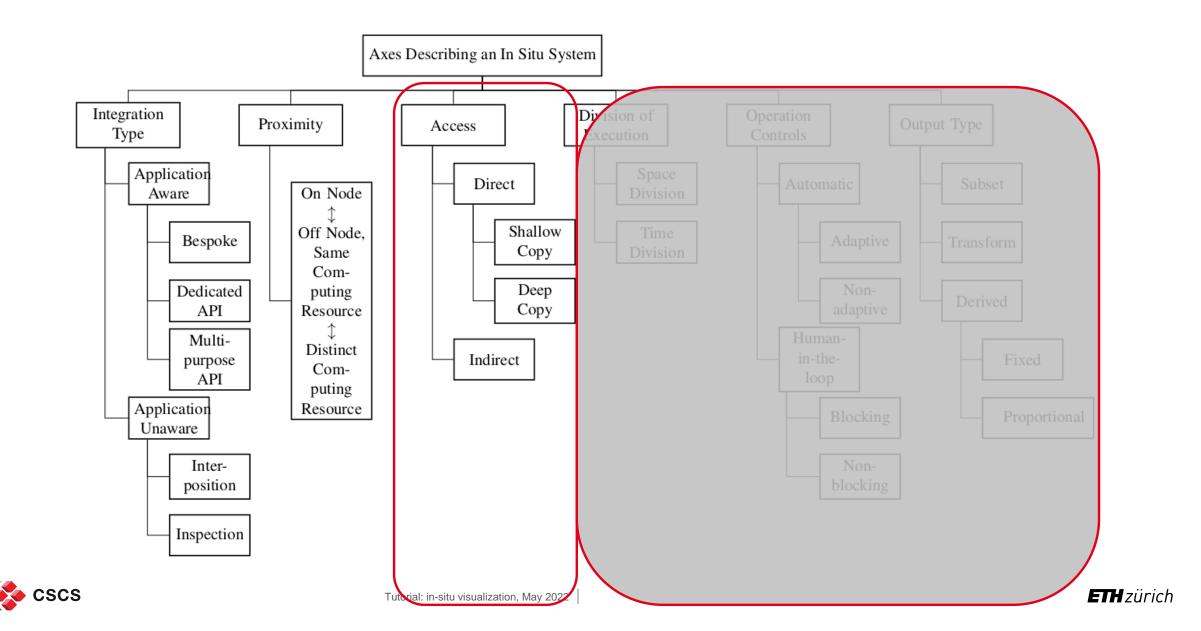
Integration Type



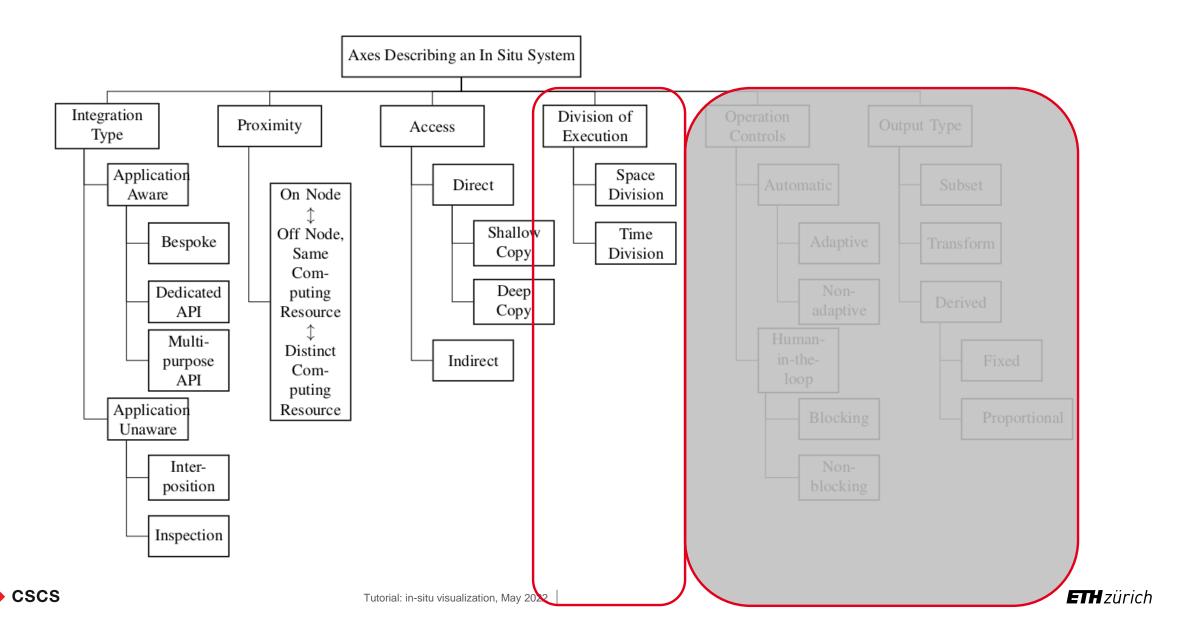
Proximity



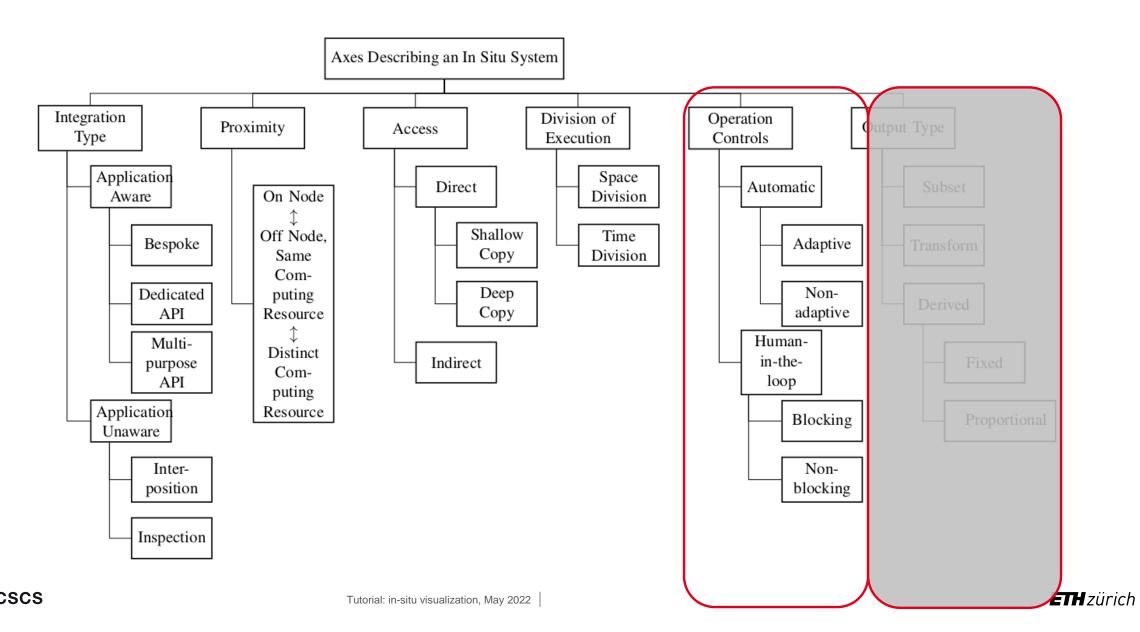
Access



Division of Execution



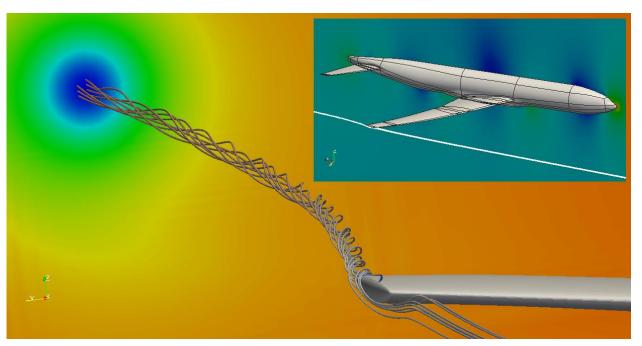
Operation Controls



Feature extraction?

What's a feature?

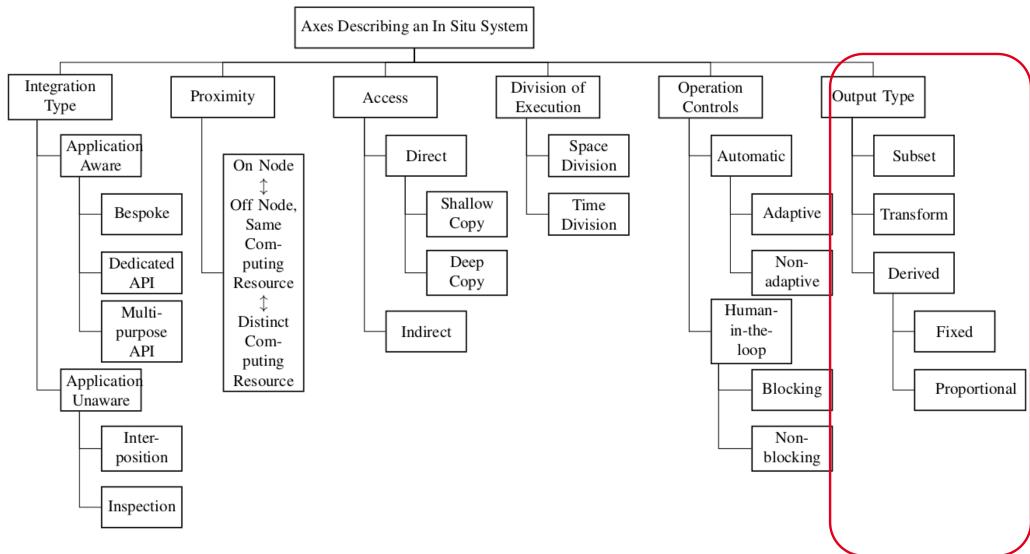
 Some small[er] data worth of interest because of domain knowledge



- Without "interactive" data exploration, it can be difficult to know a-priori which features to extract, or how to tune parameters.
- Knowing when to trigger a potentially expensive calculation would also be handy.



Output types





Outcome

- In that paper, 15 existing systems were reviewed. Half of them implement a Time Division, with On Node proximity.
 - The simulation advances, then pauses and hands control to the in situ system which completes its operations, hands control back, and so on
 - For example: Ascent, ParaView-Catalyst, Visit/Libsim
- Other examples, ADIOS, SENSEI, which can operate in different modes, sending also their data to distinct vis resources

- ADIOS, for example, can integrate with other workflow or data analytics systems without detailed knowledge of the underlying software and hardware stack
- ADIOS allows users to combine data storage, data staging, data compression, and/or data reduction (ZFP, SZ, BZip2 (compression), FlexPath, Dataspace (data staging), and coupling with ParaView, Visit, Ascent)



SENSEI

SENSEI provides a generic API to enable a "write-once, run-anywhere" environment. This approach focuses on data proximity and portability, runtime selection between running On Node or Off Node

SENSEI provides access to a diverse set of in situ analysis back-ends and transport layers such as ADIOS, Libsim, Ascent, Catalyst etc, through a simple API and data model.

Simulations instrumented with the SENSEI API can process data using any of these backends interchangeably. The back-ends are selected and configured at run-time via an XML configuration file.

For more details, refer to the SC2021 tutorial notes



What about my [the solver] internal data model?

- Do the standard visualization applications support all the data structures I use in my code?
 - => Use Data Adaptors
- Are our standard visualization applications ready to handle new alternate data representations?
 - => Use Data Converters

- In recent years, we have seen new ways of thinking about data simulations (run multiphysics code, run ensembles, use ML).
- => new data, perhaps quite different from the traditional "mesh of gridded points"







Conduit: introduction

Conduit: Simplified Data Exchange for HPC Simulations

 Conduit is an open source project from Lawrence Livermore National Laboratory that provides an intuitive model for describing hierarchical scientific data in C++, C, Fortran, and Python. It is used for data coupling between packages in-core, serialization, and I/O tasks.

- Conduit provides a convention to describe computational simulation meshes.
 This is called the Mesh Blueprint.
- Illustration of Mesh Blueprint examples

 Ascent and Catalyst use Conduit for describing data and other parameters which can be communicated between a simulation and the visualization apps.



Conduit by examples

Setup from a terminal

ssh class1??@daint.cscs.ch

git clone https://github.com/jfavre/InSitu-Vis-Tutorial2022

Read and execute contents from "PizDaint_Instructions.txt"



Conduit by examples

We will use jupyter lab and our course accounts to get compute nodes on Piz Daint

Username: class1??

Password: ???????

Reservation: in_situ

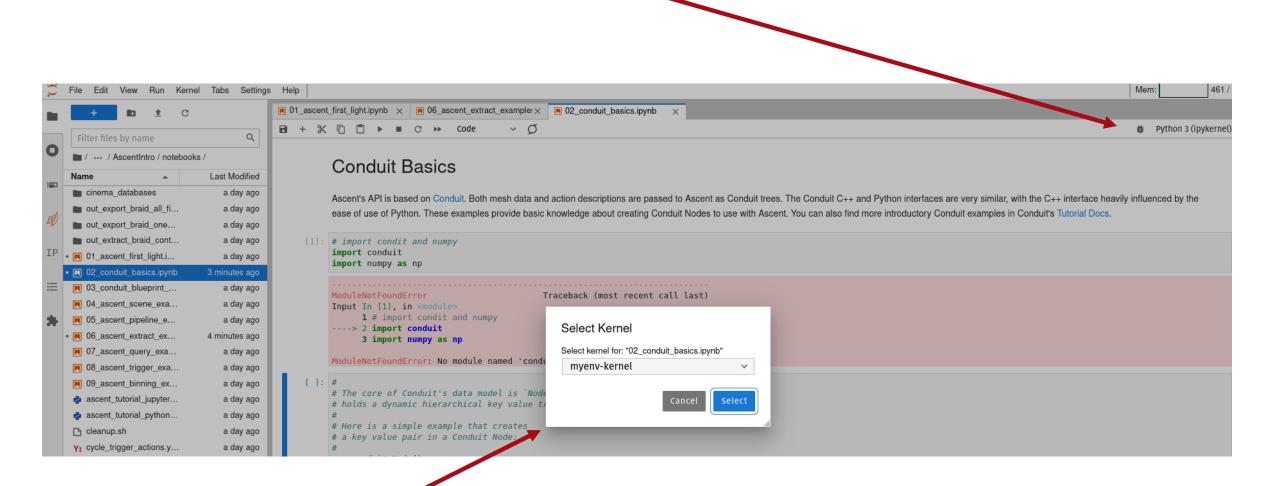
See https://jupyter.cscs.ch and get access at https://jupyter.cscs.ch

goto AscentIntro/notebooks

execute 02_conduit_basics.ipynb



import Error?



Change kernel to *myenv-kernel*







Conduit + Ascent

Ascent

Ascent is an easy-to-use flyweight in situ visualization and analysis library for HPC simulations:

- Supports: Making Pictures, Transforming Data, and Capturing Data for use outside of Ascent
- Young effort, yet already includes most common visualization operations
- Provides a simple infrastructure to integrate custom analysis
- Provides C++, C, Python, and Fortran APIs
- Ref

"The ALPINE in situ infrastructure: Ascending from the ashes of strawman", M. Larsen et al., Proc. 3rd Workshop In Situ Infrastructures Enabling Extreme Scale Anal. Vis. Denver, CO, USA, Nov. 12–17, 2017





Ascent

Ascent slides from the recent SC2021 tutorial

Thanks go to Cyrus Harrison and colleagues.



Ascent

Ascent is based on several components:

- The Conduit Mesh Blueprint!
- Runtimes providing analysis, rendering and I/O
 - Runtimes will execute a number of actions, defined by Conduit Nodes
- Data Adaptors (internal)



Ascent tutorial pages (Thanks to LLNL)

- Tutorial link
- SC2021 tutorial

docker run -p 8888:8888 -t -i alpinedav/ascent-jupyter





Ascent tutorial first example

goto AscentIntro/notebooks

Run 01_ascent_first_light.ipynb





Ascent tutorial examples,

Generating time dependent data

Run 03_conduit_blueprint_mesh_examples.ipynb

Rendering images with Scenes

Run 04_ascent_scene_examples.ipynb

Transforming data with Pipelines

Run 05_ascent_pipeline_examples.ipynb



Running with different visualization pipelines

In its basic form, we usually have a default scene description into the Ascent bridge code

- At run-time, we can supply different scenes, with additional pipelines
- If present, a file called "ascent_actions.{json,yaml}" will be read, overriding the actions previously set
- http://www.yamllint.com/ is your friend

 See example in Examples/LULESH/{ascent_actions.yaml,trigger_ascent_actions.yaml}



Let's add a scene description

```
"action": "add_scenes",
       "scenes": {
          "s1": {
             "plots": {
               "p1": { "type": "pseudocolor", "field":
"Density"} },
             "renders": {
               "r1": {
                 "image_prefix": "DensityImage.%05d",
                 "camera": {
                  "look_at": [0, 0, 0],
                  "position": [-2.17, 1.79, 1.80],
                  "up": [0.44, 0.84, -0.30]
```

```
action: "add_scenes"
scenes:
 s1:
  plots:
    p1:
    type: "pseudocolor"
    field: "Density"
renders:
    r1:
     image_prefix: "DensityImage.%05d"
     camera:
      azimuth: 30
      elevation: 11
```

Let's add a pipeline description

```
"action": "add_pipelines",
       "pipelines": {
          "pl1": {
            "f1": {
               "type": "threshold",
               "params": {
                  "field": "Density",
                  "min_value": 1.4,
                  "max_value": 2000
```



The scene description is refined with the new pipeline

```
"action": "add_scenes",
       "scenes": {
          "s1": {
            "plots": {
               "p1": {
                  "type": "pseudocolor",
                  "pipeline": "pl1",
                  "field": "Density"
            },
```



Example: Instrument an SPH simulation package with Ascent

 The smooth particle hydrodynamics (SPH) technique is a purely Lagrangian method. SPH discretizes a fluid in a series of interpolation points whose distribution follows the mass density of the fluid.

- PASC, the Swiss Platform for Advanced Scientific Computing initiative, supports the SPH-EXA project developing an SPH library.
- SPH-EXA is a C++17 headers-only code with no external software dependencies. The parallelism is currently expressed via the following models: MPI, OpenMP, CUDA and HIP.



Instrument the SPH-EXA simulation package with Ascent

Define a Conduit mesh definition

Define a Conduit <u>scene</u> definition

About 150 lines of code. Total!





Using Conduit, a particle set is trivially described [the coordinates]

```
particle_set = """
```

coordsets:

coords:

type: "explicit"

values:

x: [0.0, 10.0, 20.0, 30.0]

y: [0.0, 10.0, 20.0, 30.0]

z: [0.0, 10.0, 20.0, 30.0]

11 11 11

```
conduit::Node mesh;
mesh["state/cycle"].set_external(&d.iteration);
mesh["state/time"].set_external(&d.ttot);
mesh["coordsets/coords/type"] = "explicit";
mesh["coordsets/coords/values/x"].set_external(&d.x);
mesh["coordsets/coords/values/y"].set_external(&d.y);
mesh["coordsets/coords/values/z"].set_external(&d.z);
// The heavy-data is available via shallow-copy links
```



Using Conduit, a particle set is trivially described [the topology]

```
particle_set = """
 topologies:
   mesh:
    type: "unstructured"
    elements:
    shape: "point"
    connectivity: [0, 1, 2, 3]
    coordset: "coords"
11 11 11
```

```
mesh["topologies/mesh/type"].set("unstructured");
mesh["topologies/mesh/elements/shape"].set("point");
mesh["topologies/mesh/coordset"].set("coords");
std::vector<int> conn(N); // N is # of particles
std::iota(conn.begin(), conn.end(), 0);
mesh["topologies/mesh/elements/connectivity"].set(conn);
```



Using Conduit, a particle set is trivially described [the solution fields]

```
particle_set = """
fields:
 rho:
  association: "vertex"
  values: [-1, -2, -3, -4]
  topology: "mesh"
  volume_dependent: "false"
  units: "g/cc"
11 11 11
```

```
auto fields = mesh["fields"];
// Density scalar field
 fields["rho/association"].set("vertex");
fields["rho/topology"].set("mesh");
 fields["rho/volume_dependent"].set("false");
// Conduit supports shallow copy
 fields["rho/values"].set_external(&d.rho);
```



Use one of conduit's relay protocol to save data to disk?

c++: extract node from SPH-EXA In Python import conduit.relay as relay if relay.io.about()["protocols/hdf5"] == "enabled": relay.io.save(mesh, "/dev/shm/foo.hdf5")

```
h5ls -r /dev/shm/foo.hdf5
               Group
/coordsets
                   Group
/coordsets/coords
                     Group
coordsets/coords/type Dataset {9}
coordsets/coords/values Group
/coordsets/coords/values/x Dataset {4}
/fields
                Group
/fields/rho
                 Group
/fields/rho/values
                    Dataset {4}
/topologies
                  Group
/topologies/mesh
                     Group
topologies/mesh/coordset Dataset {7}
/topologies/mesh/elements Group
/topologies/mesh/elements/connectivity Dataset {4}
```





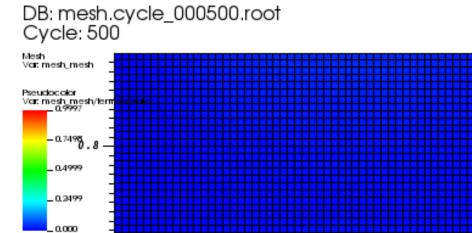


Supplementary material

Instrument the Jacobi-Python example with Ascent

Define a Conduit mesh definition

Define a Conduit <u>scene</u> definition



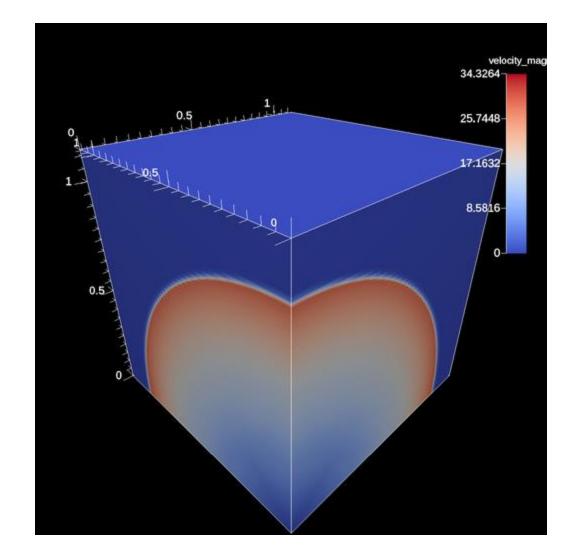
Max: 0.9997 0 . 6 Mrx: 0.000



Instrument the LULESH example with Ascent

Define a Conduit <u>mesh</u> definition

Define a Conduit <u>scene</u> definition











Coffee Break





Before jumping to ParaView Catalyst, let's first introduce ParaView

ISC 22 - Catalyst 2

François Mazen



ParaView

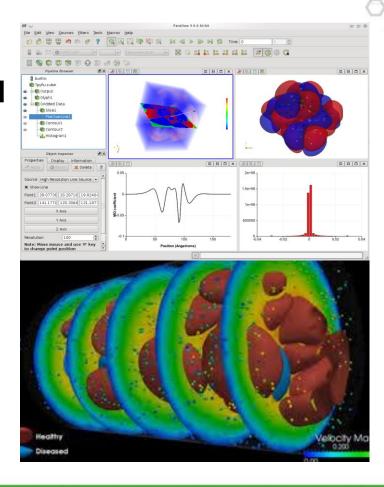


ParaView / High-Performance Post-Processing

- Open-source, multi-platform, data analysis and visualization application
- Analysis of extremely large datasets using distributed memory computing resources
- Over 20 years of development
- Contributions from over 270 developers
- Over 1.6 million lines of code
- Over 150k yearly downloads
- www.paraview.org

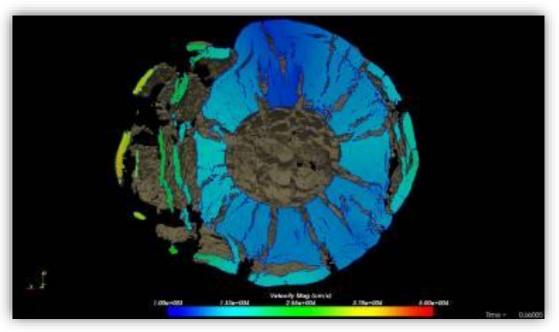




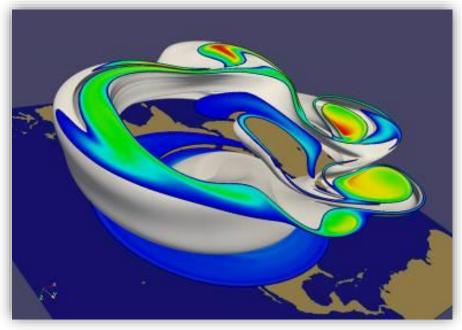


Extremely Large Data

1 billion cell asteroid detonation simulation



1/2 billion cell weather simulation



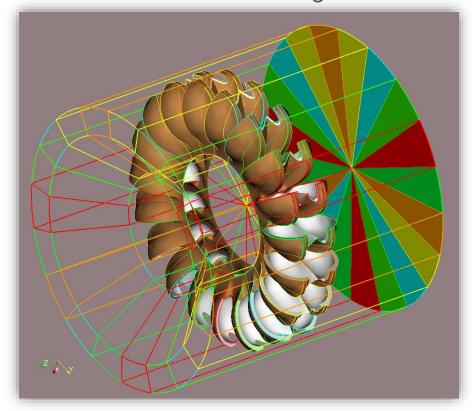
Source: Sandia National Lab

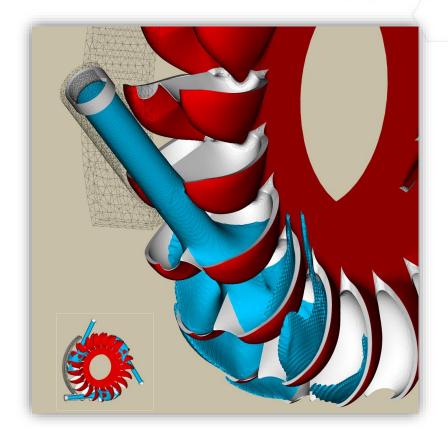




Fast Large Data Interaction

CFD simulation of 20-30 million cells with load balancing

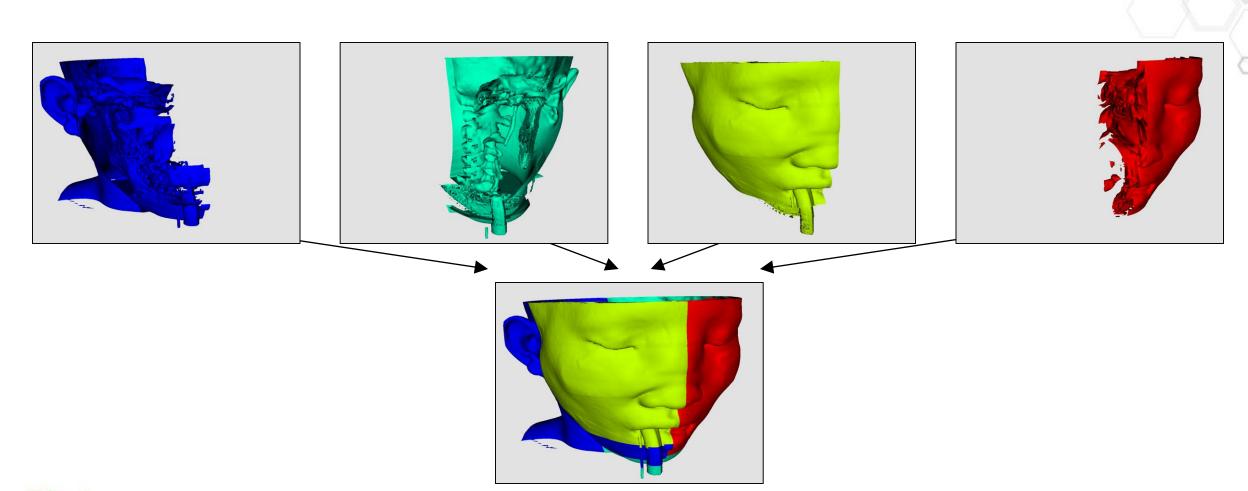




Source: Swiss National Supercomputing center

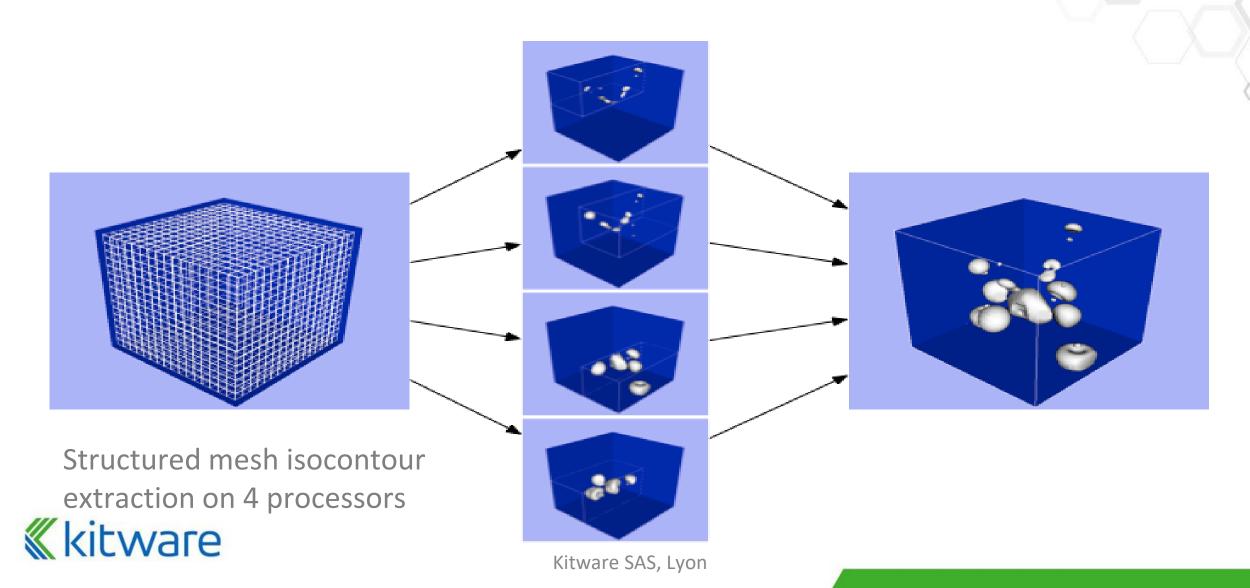


Parallel Rendering

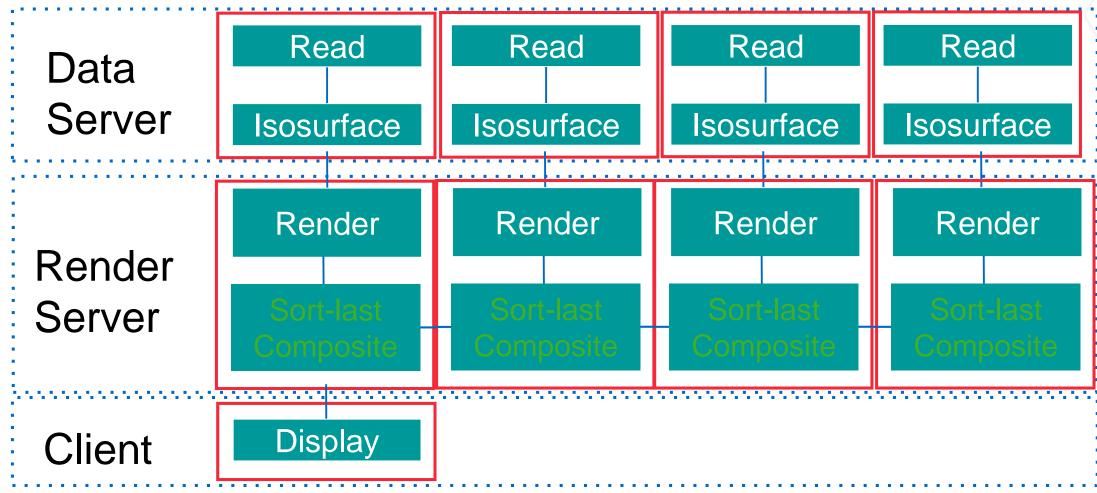




Distributed Processing



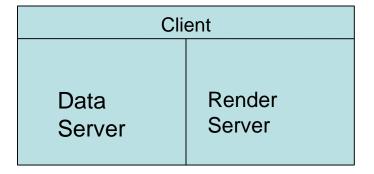
Data Parallel Pipelines



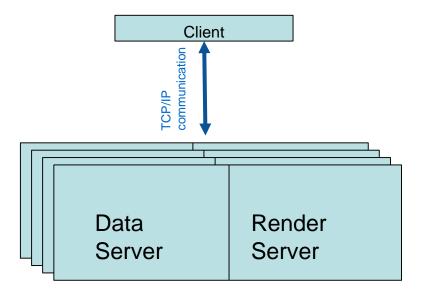


Deployment Topologies

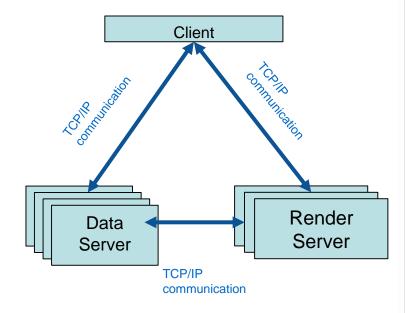
Builtin Server



Client-Server

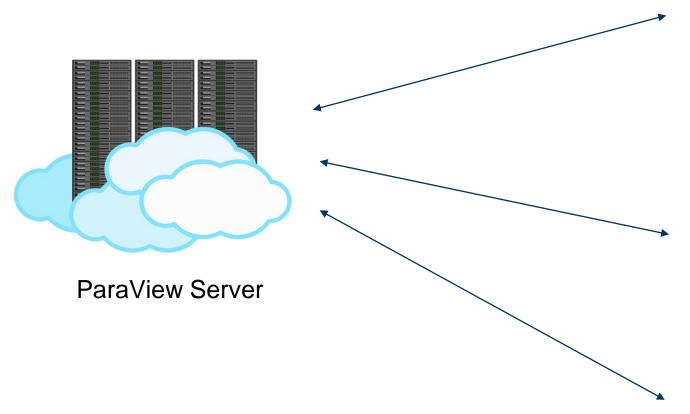


Client-Render Server-Data Server



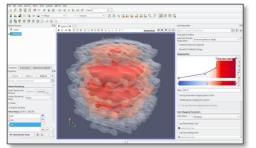


Running ParaView





Batch mode (ParaView Python)



ParaView Client





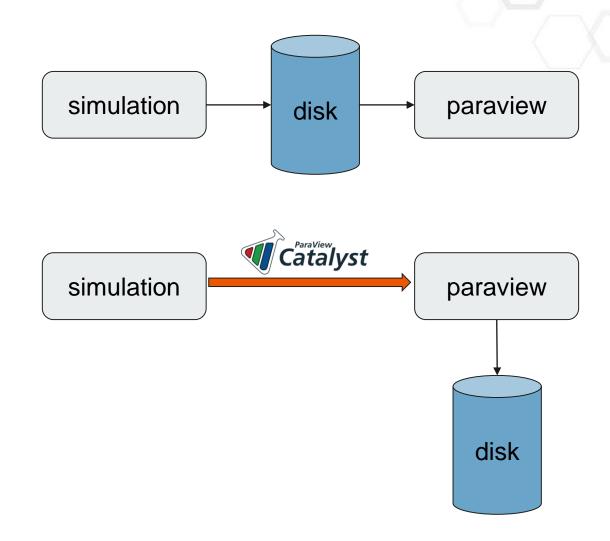
In-Situ Analysis



What is ParaView Catalyst?

- An in-situ framework embedded with ParaView
 - in-situ analysis: uses ParaView analysis capabilities
 - live visualization: uses ParaView to connect to the simulation
 - easy configuration: uses ParaView to generate scripts
 - open source

- Challenges : explosion of scientific data
 - Storage
 - Bandwidth





Saving Data Storage

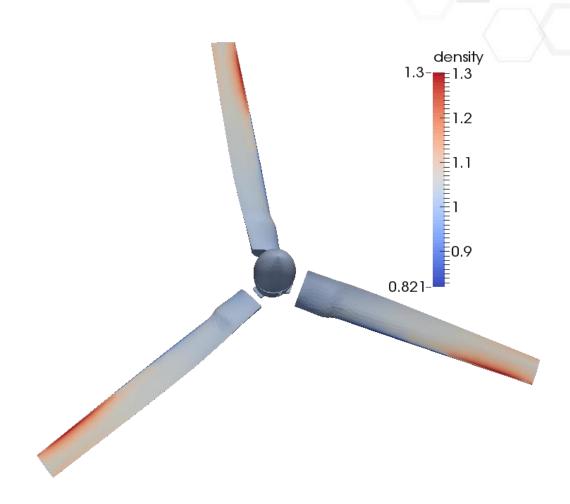
Rotorcraft simulation output size for a single time step:

Full data set – 448 MB

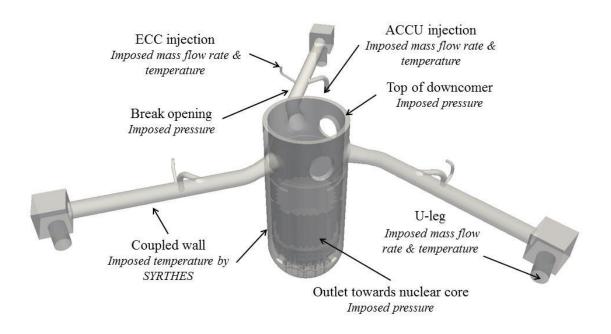
Surface of blades – 2.8 MB

Image – 71 KB

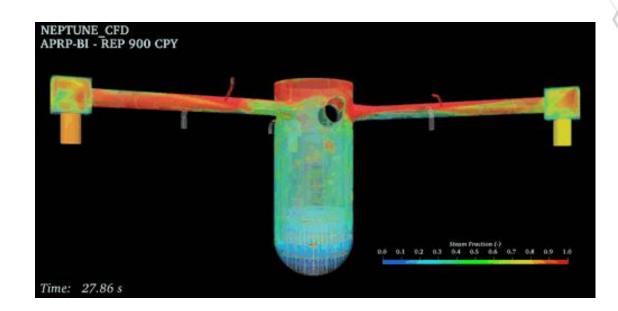




ParaView Catalyst in Code_Saturne Example



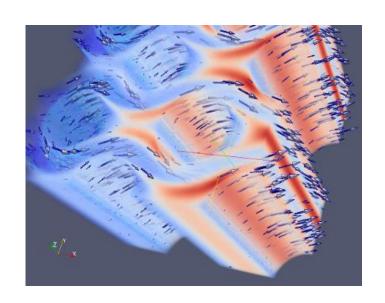
- Fluid mesh: about 6 500 000 cells
- Solid mesh: around 4 000 000 cells



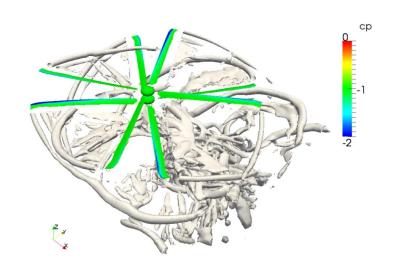
Courtesy of Nicolas MERIGOUX & Yvan FOURNIER, EDF R&D



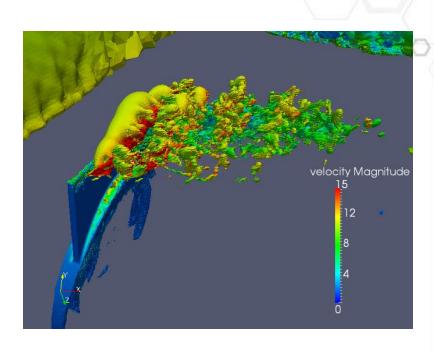
More Catalyst Examples



Code Saturne (EDF)



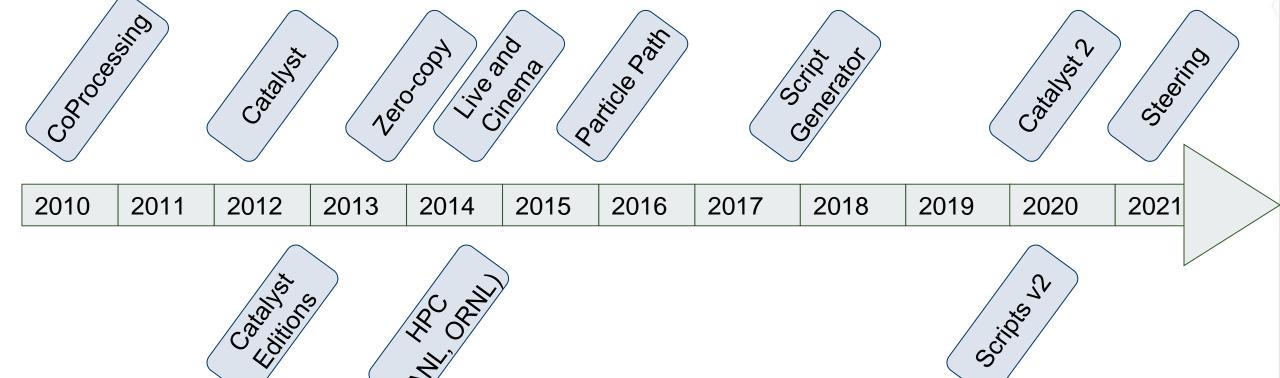
Helios (Army Aeroflighdynamics Directorate)



Phasta (UC-Boulder)



Catalyst History



My Selly



Catalyst 2



Catalyst 2 : challenges

Make it easy to develop

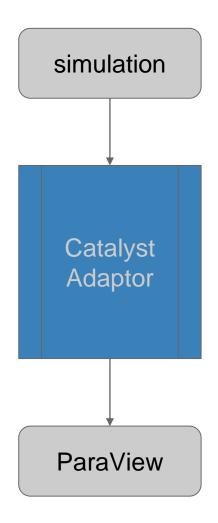
Make it easy to build

Make it easy to deploy

Make it easy to maintain and upgrade



Catalyst 2 : easy development



- Avoid need to understand VTK data model
- Provide mechanism to provide data with zerocopy & meta-data to interpret it

→ Conduit!





Catalyst 2: simple API and stable API

https://catalyst-in-situ.readthedocs.io

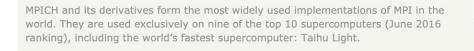
```
enum catalyst_status catalyst_initialize(const conduit_node* params);
enum catalyst_status catalyst_finalize(const conduit_node* params);
enum catalyst_status catalyst_execute(const conduit_node* params);
enum catalyst_status catalyst_results(conduit_node* params);
enum catalyst_status catalyst_about(conduit_node* params);
```



Catalyst 2 API

 Inspired by "MPICH ABI compatibility initiative" (https://www.mpich.org/abi/)





Interface (MPI) standard.





Catalyst 2 implementations

Change implementation at run-time

- Environment variables
 - CATALYST_IMPLEMENTATION_PATHS
 - CATALYST_IMPLEMENTATION_NAME
- Dedicated keys in conduit node of catalyst_initialize
 - catalyst_load/search_paths
 - catalyst_load/implementation
- Legacy way: LD_LIBRARY_PATH

Conduit Node
Mesh Blueprint

VTK Data
Object

vtkDataObjectToConduit

vtkConduitSource

Several implementations

- catalyst-stub: no external dependencies, suitable for building
- catalyst-paraview: uses ParaView and VTK to work on data
- Debug, replay, etc.



Catalyst 2 Steering

Drive the simulation from ParaView via Catalyst API and Live Visualization

```
enum catalyst_status
catalyst_results(conduit_node* params);
```





Catalyst 2 : goals achieved!

- Easy to develop
 - No need to understand VTK
 - No data conversion

- Easy to build
 - Few dependencies using catalyst-stub
 - No CMake

Easy to deploy

- Several catalyst implementations deployment
- Not link to specific ParaView version

- Easy to maintain
 - Stable C API/ABI
 - No need to rebuild simulation for new ParaView version



Catalyst Road Map

- Meta data on meshes
 - Defined in the mesh/state/my_meta_data conduit node
 - Becomes VTK Field Data

- In-transit analysis with Catalyst
 - Based on ADIOS







Conduit + ParaView Catalyst

ParaView Catalyst practical how-to

- Instrumented our simulation code? Yes!
- Compiled with ParaView Catalyst? Yes!

How do we run and specify visualization tasks????





The ParaView Extractors

- Extractors are items in the visualization pipeline that can save data or images at a user-chosen temporal frequency.
 - Data Extractors
 - Image Extractors
- Using Extractors, no custom code per iteration is really necessary in the majority of cases. One can simply use Extractors to save out images from views or data extracts from filters and other data producers. (*more on that later*)



The ParaView Catalyst Python scripts

- Python scripts, are written by the ParaView application, given a representative template input file, and a set of visualization filters interactively tuned by the user in an offline fashion [not connected to a running solver]
- We create a visualization pipeline with the numerous visualization filters and rendering options available in ParaView, and we add "Extractors"
- ParaView provides hybrid parallelism out-of-the-box
 - MPI-enabled
 - TBB multi-threading
 - CUDA-enabled filters
- Some [or most of the] ParaView visualization code will be tightly integrated with the solver memory and execution space.



ParaView Catalyst Python scripts

- Can be generated from an interactive ParaView session, and later fine-tuned
- Are completely interchangeable between the batch-mode ParaView execution (reading data from disk), and the in-situ execution





Catalyst scripts (batch-mode

VS.

in-situ mode)

```
grid = OpenDataFile(registrationName='grid', filename=['/LULESH/datasets/data_000009.vtpd'])
```

renderView1 = GetRenderView()

rep = Show()

ColorBy(rep, ['POINTS', 'velocity'])

Render()

execute in batch-mode with data read from disk

from paraview.simple import SaveExtractsUsingCatalystOptions

SaveExtractsUsingCatalystOptions(options)

grid = TrivialProducer(registrationName='grid')

renderView1 = GetRenderView()

rep = Show()

ColorBy(rep, ['POINTS', 'velocity'])

v = CreateExtractor('VTPD', grid)

v.Trigger = 'TimeStep'

v.Trigger.Frequency = 30

v.Writer.FileName = 'data_{timestep:06d}.vtpd'

ParaView-Catalyst Blueprint

The Protocol is rather simple:

 Defines the options accepted by catalyst_initialize(); these include things like ParaView Python scripts to load, directories to save data

 Defines the protocol for catalyst_execute() and includes information about Catalyst channels i.e. ports on which data is made available to in situ processing as well as the actual data from the simulation

Defines the protocol for catalyst_finalize()



ParaView-Catalyst Blueprint

Similar to the Conduit Mesh Blueprint.

```
node["catalyst/scripts/script/filename"] =
```

```
node["catalyst/state/cycle"] =
node["catalyst/state/time"] =
```

```
node["catalyst/channels/grid/type"] = "mesh"
node["catalyst/channels/grid/data"] =
```



Code instrumentation -

The Catalyst glue code for the SPH-EXA solver is 144 lines of code

Enabling in-situ visualization can be optionally compiled

before

```
int main(int argc, char** argv)
 MPI_Init_and_Code_Init();
 for (d.iteration = 0; d.iteration <= maxStep; d.iteration++)
 Solve_For_Each_Timestep();
 return exitSuccess();
```



Code instrumentation -

- The Catalyst glue code for the SPH-EXA solver is 144 lines of code
- The execution driver is instrumented with 4 lines of code

Total: 148 lines of code

after

```
#include "CatalystAdaptor.h"
int main(int argc, char** argv)
 MPI_Init_and_Code_Init();
 CatalystAdaptor::Initialize(argc, argv);
 for (d.iteration = 0; d.iteration <= maxStep; d.iteration++)
 Solve_For_Each_Timestep();
 CatalystAdaptor::Execute(d, domain.startIndex());
 CatalystAdaptor::Finalize();
 return exitSuccess();
```

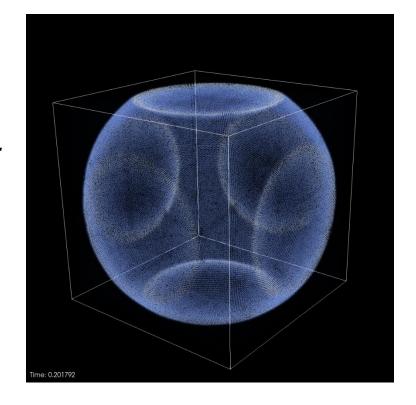


What about "Operations Control"?

Adaptive Control

- Choose different pathes of execution based on queries/triggers
- The catalyst_execute(ConduitNode) script can be customized

```
def catalyst_execute(node):
  threshold = 1.4
  if reader.PointData["Density"].GetRange()[1] > threshold:
    print("density at timestep", node.timestep)
    # Extract particles where Density > threshold
  else:
    # use ALL particles
```





What about "Operations Control"?

Human in the loop

ParaView has a very efficient client-server mode for post-hoc visualization

- Uses a powerful, remote server to do the heavy work:
 - Parallel I/O
 - Parallel filtering
 - Parallel Image Composition
- Has a very efficient, threshold-based, Image Delivery engine
- The results of the different visualization operations stay on the server (Mbytes, Gbytes, Tbytes (choose your flavour))
- Only the image gets sent over the network to the client (1024x1024 pixels, RGB).



ParaView-Catalyst uses the same engine

- Open a port on a reverse tunnel to my desktop
 - ssh -R 22222:localhost:22222 daint103.cscs.ch
 - Tell my desktop client to send a request for connection on that port
 - my code instrumentation listens to incoming calls with four additional lines of Python code:

```
from paraview import catalyst
options = catalyst.Options()
options.EnableCatalystLive = 1
options.CatalystLiveURL = 'daint103:22222'
```

Demonstration of Catalyst Live



Other examples: LULESH instrumented for ParaView Catalyst

ASC Proxy Apps

Instrumented with the <u>VTK-based</u> Catalyst Instrumented with the <u>Conduit-based</u> Catalyst

Thanks to Utkarsh Ayachit (Kitware)





```
renderView1 = CreateView('RenderView')
selection=SelectPoints()
selection.QueryString="Density >= 1.4"
extractSelection = ExtractSelection()
thresholdDisplay = Show(extractSelection)
ColorBy(thresholdDisplay, ['POINTS', 'Density'])
pNG1 = CreateExtractor('PNG', renderView1)
pNG1.Trigger = 'TimeStep'
pNG1.Writer.FileName =
'threshold {timestep:06d}{camera}.png'
pNG1.Trigger.Frequency = 100
```

```
"action": "add_pipelines",
      "pipelines": {
         "pl1": {
            "f1": {
               "type": "threshold",
[...]
"action": "add scenes",
       "scenes": {
         "s1": {
            "plots": {
               "p1": {
                 "type": "pseudocolor",
                 "pipeline": "pl1",
[...]
       "renders": {
               "r1": {
                "image prefix": "ThresholdImage.%05d",
```





Third Method. Use Ascent and ParaView Python scripts

Ascent, passing data to a ParaView script

Ascent Documentation

- The instrumented simulation sends a tree structure (json like) that describes the simulation data using the Conduit Blueprint Mesh specification.
- A ParaView plugin constructs one of the following datasets: vtkImageData, vtkRectilinearGrid, vtkStructuredGrid or vtkUnstructuredGrid
- This data is converted to a VTK format using shallow copies for data arrays



Ascent, passing data to a ParaView script

Ascent's "action" node is:

```
"action": "add_extracts",
    "extracts":
    {
        "e1":
        {
            "type": "python",
            "params":
            {
                 "file": "paraview-vis.py"
            }
        }
    }
}
```

 The ParaView script describes a "standard" ParaView pipeline and can save images and any other derived data to disk



Summary

- Adopting the two-Conduit-based environments presented is rather simple, very intuitive. The bridges to ParaView or Ascent are now hiding all the complexity from the code developers
- Both environments have proven tracks of high-scale deployments
- What's next:
 - evaluate performance, memory consumption,
 - visualization pipeline creation and tuning, and re-usability
- For the first time in several decades of developing data formats plugins (converters, adaptors, ..), I don't have to worry about this anymore, and can concentrate on what I love to do best: !! Visualization!!
- Already 10 years ago, while presenting libSim, people would tell me

"we're **very** tight on memory"

Adding an in-transit layer is the next venue to explore. Appointment for a CSCS tutorial in 2023.



Resources

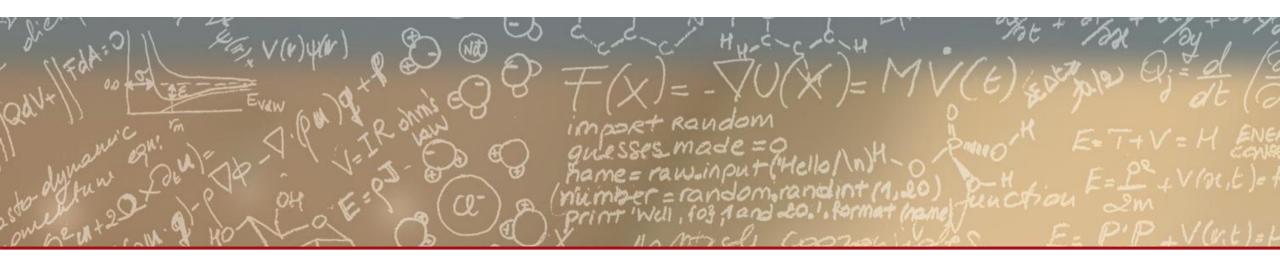
- https://github.com/jfavre/InSitu-Vis-Tutorial2022
- My slides
- https://dav.lbl.gov/events/SC21_SENSEI_Ascent_Tutorial/











Thank you for your attention.